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Handling Student
Mobility in Educational
Research: Practical
Examples from Maryland

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#### Overview

- Introduction and background
- Rates of student mobility in Maryland
  - Student characteristics
  - School characteristics
- Approaches for handling student mobility in educational research
- Summary
- Questions



## Introduction and Background



#### Introduction (1)

- Researchers using state longitudinal data systems are often interested in examining predictors of student outcomes
  - What is the relation between early disadvantage and later academic achievement?
- Predictors of interest may be at the student- or the school-levels
  - Student-level: student disadvantage
  - School-level: school concentration of disadvantage
- What % of the variance in student outcomes is due to the school level?

25-60%



#### Introduction (2)

- Education data are inherently clustered (e.g., students are nested within schools)
- Analyzing predictors at one level without the other will produce misleading results
- Hierarchical Linear Modeling (HLM; Raudenbush & Bryk, 2002) is the traditional statistical approach for correctly adjusting for clustering of students within schools
- However, analyses become complicated when students attend more than one school over time (i.e., the student mobility problem)



#### The Student Mobility Problem

- Student mobility when students change schools either within the academic year or between academic years (Rumberger, 2015)
- Mobility rates are high in the United States
  - Estimates range from 15% 45% of students
  - Varies by student subgroup
  - Higher rates for minority students, low-income students, and lower performing students
  - Higher rates in urban schools and lower performing schools

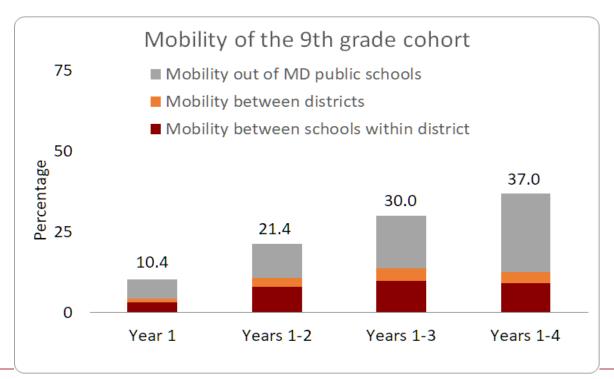
(Rickles et al., 2018; U.S. Government Accounting Office, 1994)



# Rates of Student Mobility in Maryland



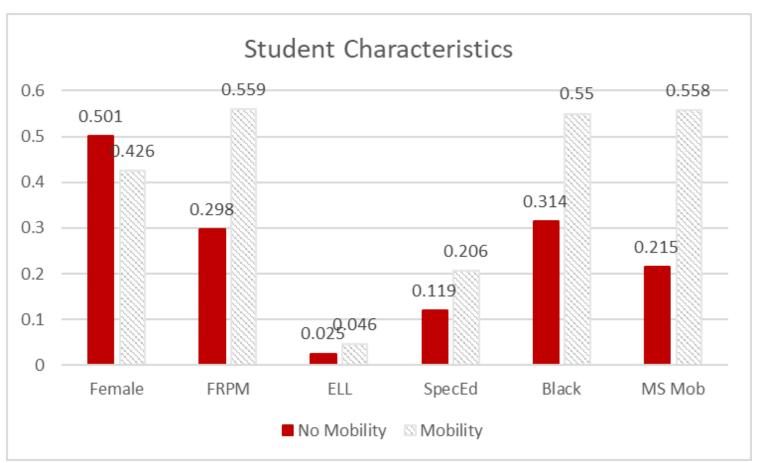
#### Student Mobility Rates in Maryland (1)



- By the end of 4 years, 37% of the 9th grade cohort experienced mobility out of the school where they started 9th grade.
- Most of this mobility was out of MD public schools altogether.

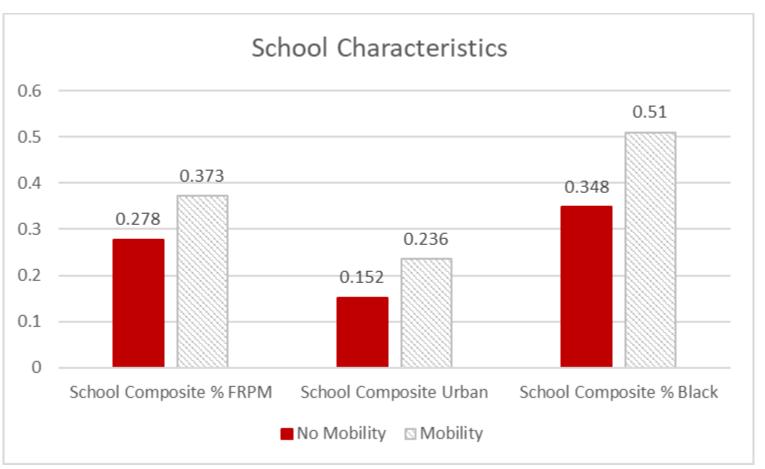


#### Student Mobility Rates in Maryland (2)





#### Student Mobility Rates in Maryland (3)





## Questions?

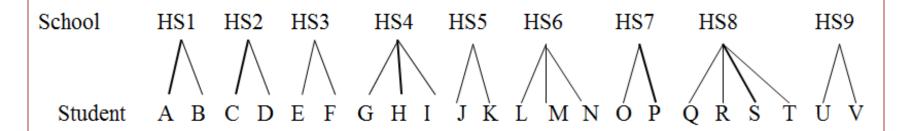


# Approaches for Handling Student Mobility in Educational Research

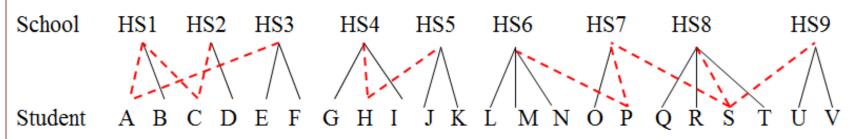


#### Hierarchical Linear Modeling (HLM)

HLM is appropriate when each student is nested within *only one* school (HS=high school):



... but real-world data aren't purely hierarchical!



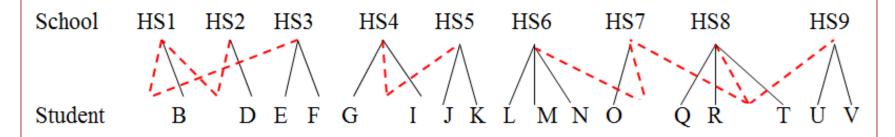


#### Comparison of Three Approaches

- HLM with deletion
- HLM with first school assigned
- Multiple membership modeling



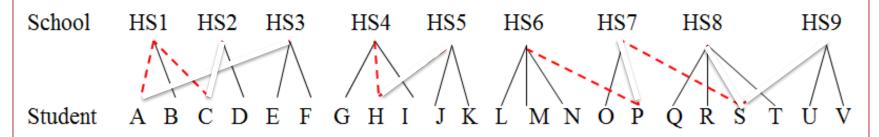
#### **HLM** with Deletion



- Deletes mobile students before conducting statistical analyses
- Reduces statistical power (fewer students in the sample)
- Limits external validity (generalizability) b/c now the sample is only representative of non-mobile students



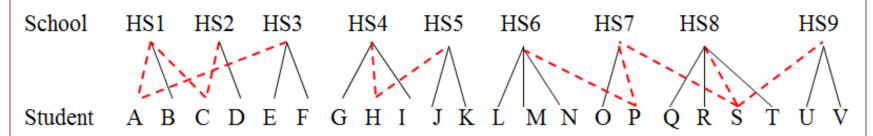
#### **HLM** with First School Assigned



- Assigns mobile students to their first school attended before conducting statistical analyses
- Reduces internal validity
- May lead to misattributing some of the school-level variance to the student-level (see Chung & Beretvas, 2012)



#### Multiple Membership Modeling



- Accounts for each school attended by mobile students by creating weights for each school
- Example:
  - Equal Weighting Student A attends HS1 and HS3;
     each school is weighted 0.50
  - Proportional Weighting—Student A attends HS1 for 75% of the year and HS3 for 25% of the year; HS1 is weighted 0.75 and HS3 is weighted 0.25



## Questions?



# Comparing Approaches for Handling Student Mobility

What is the relation between early disadvantage and later academic achievement?

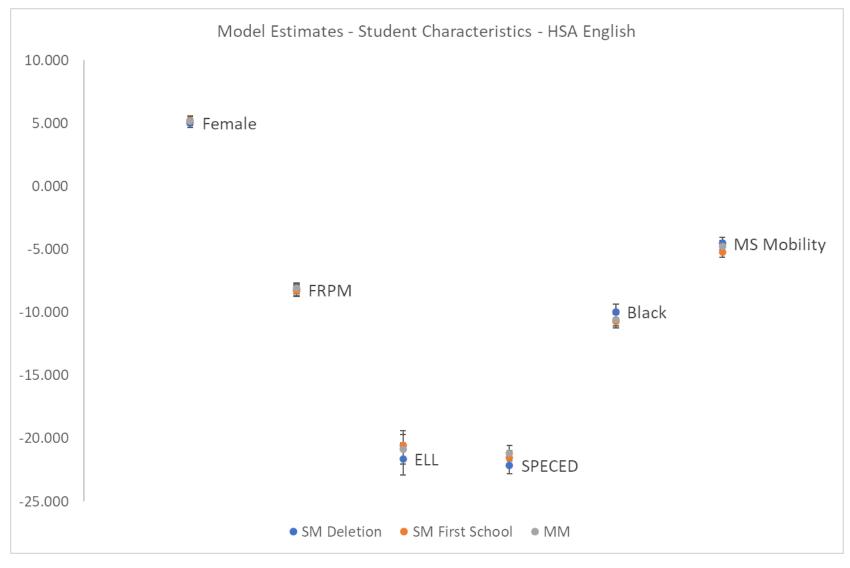
- → Traditional HLM model delete mobile students
- → Traditional HLM model assign mobile students to their first school
- → Multiple membership (MM) model
- → Compare the results



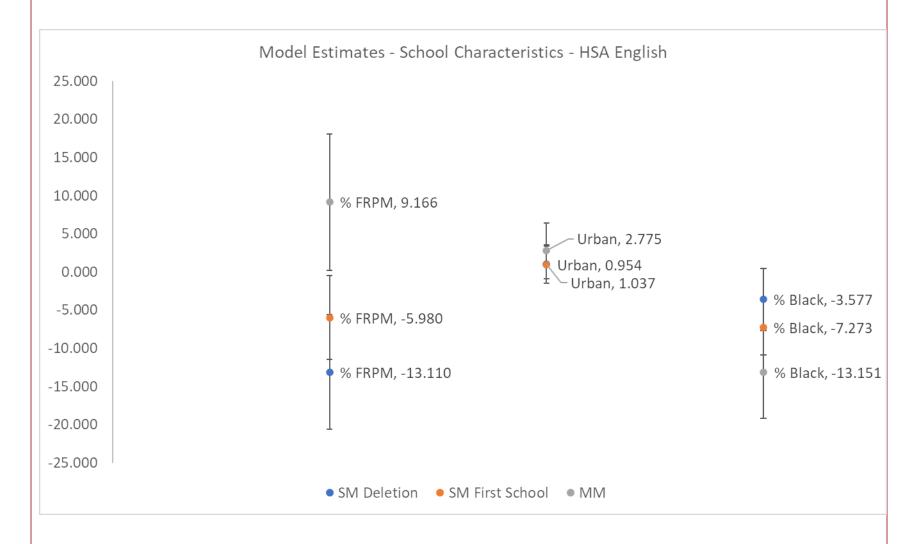
#### HSA English Models (N and Variance)

	SM Deletion	SM First School	MM
N Level 1 (student)	53,954	63,011	63,011
N Level 2 (school)	225	265	265
ICC	0.29	0.30	0.47







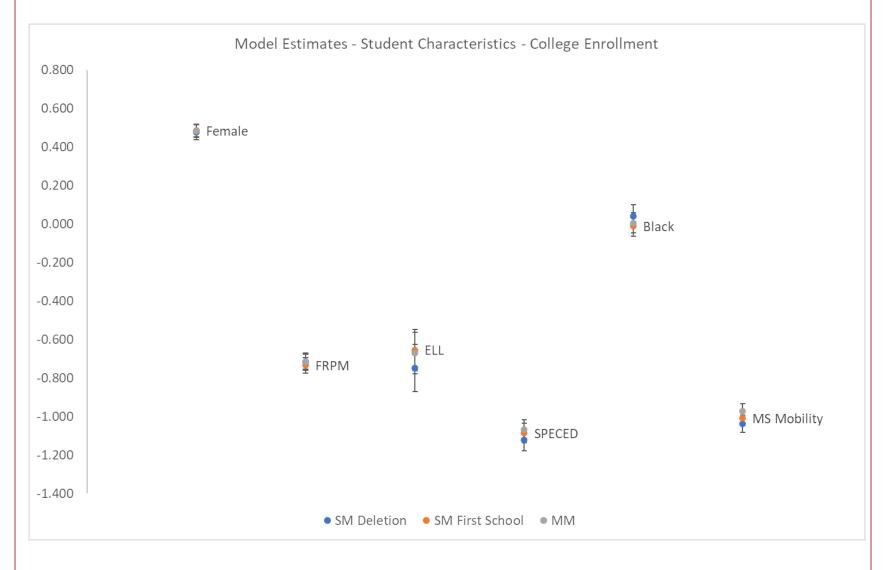




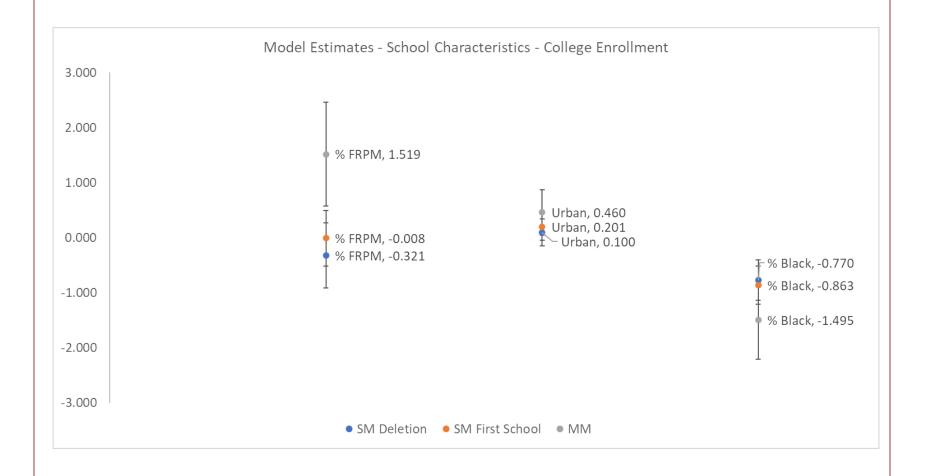
# College Enrollment Models (N and Variance)

	SM Deletion	SM First School	MM
N Level 1 (student)	59,774	75,279	75,279
N Level 2 (school)	253	307	307
ICC	0.31	0.33	0.63











## Questions?



## Summary



#### Summary

- The choice of modeling approach matters for substantive interpretation and subsequent policy decisions
- The loss of students and schools when ignoring student mobility results in threats to external validity
  - Deleting mobile students results in disproportionate losses of some types of students (EL, minority, FRPM)
- Assigning students to their first school attended may result in threats to internal validity
  - Especially when interested in school-level predictors
- Using multiple membership modeling represents a viable solution for handling student mobility



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#### For More Information

MLDS Center website

https://mldscenter.maryland.gov/

Working paper available upon request – Email:

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